



# Control Systems

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## Part 2: Modeling of Dynamic Systems



## Learning objectives

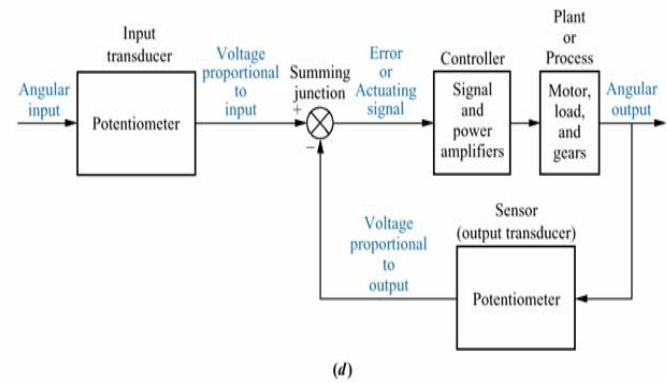
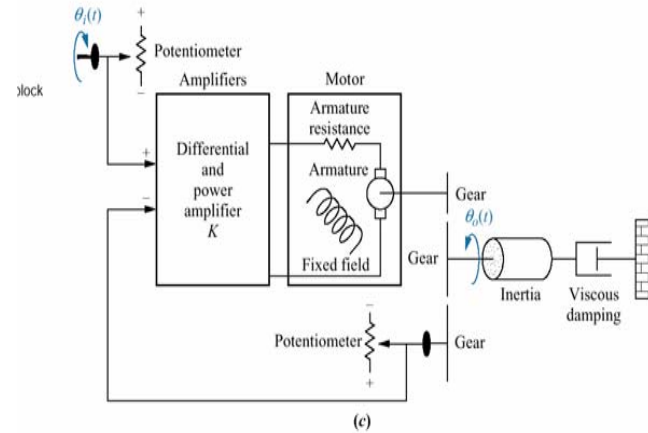
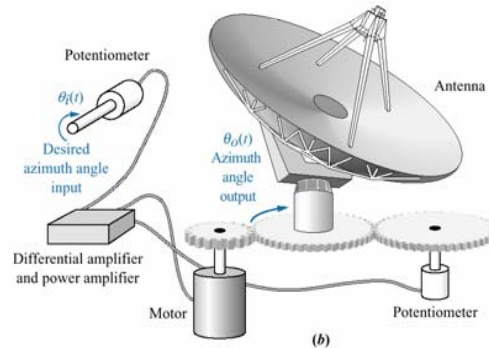
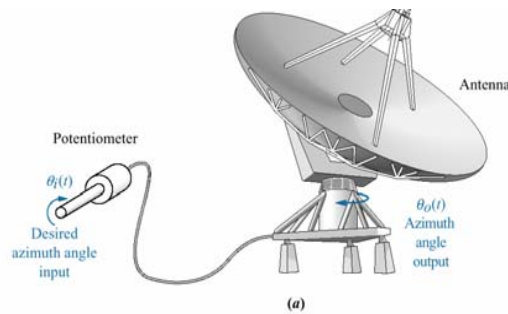
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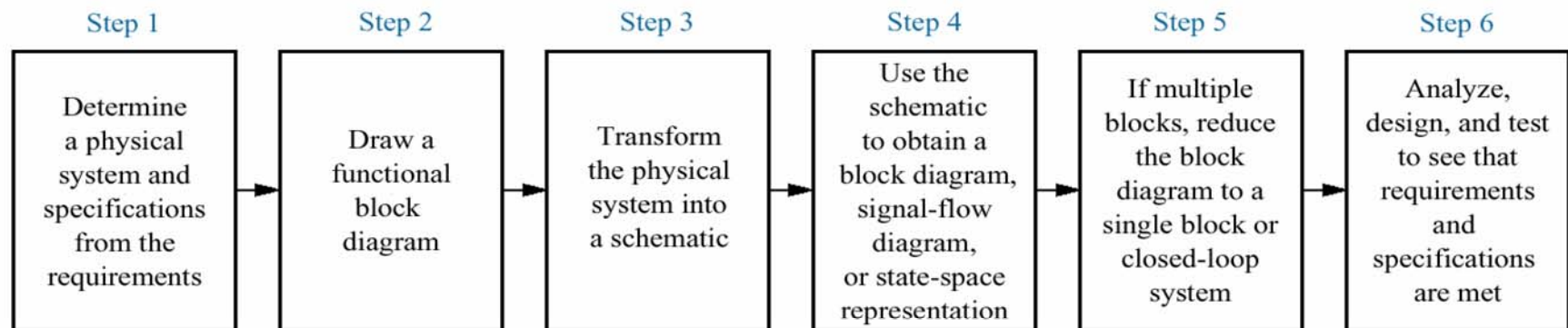


- To derive mathematical representation of physical systems
- To solve differential equations for time domain responses
- To convert the differential equations to another domain, i.e. Laplace Transforms
- To represent system in terms of transfer functions using Laplace transforms.

# Mathematical representation of the system

- (a) system concept;
- (b) detailed layout;
- (c) schematic;
- (d) functional block diagram



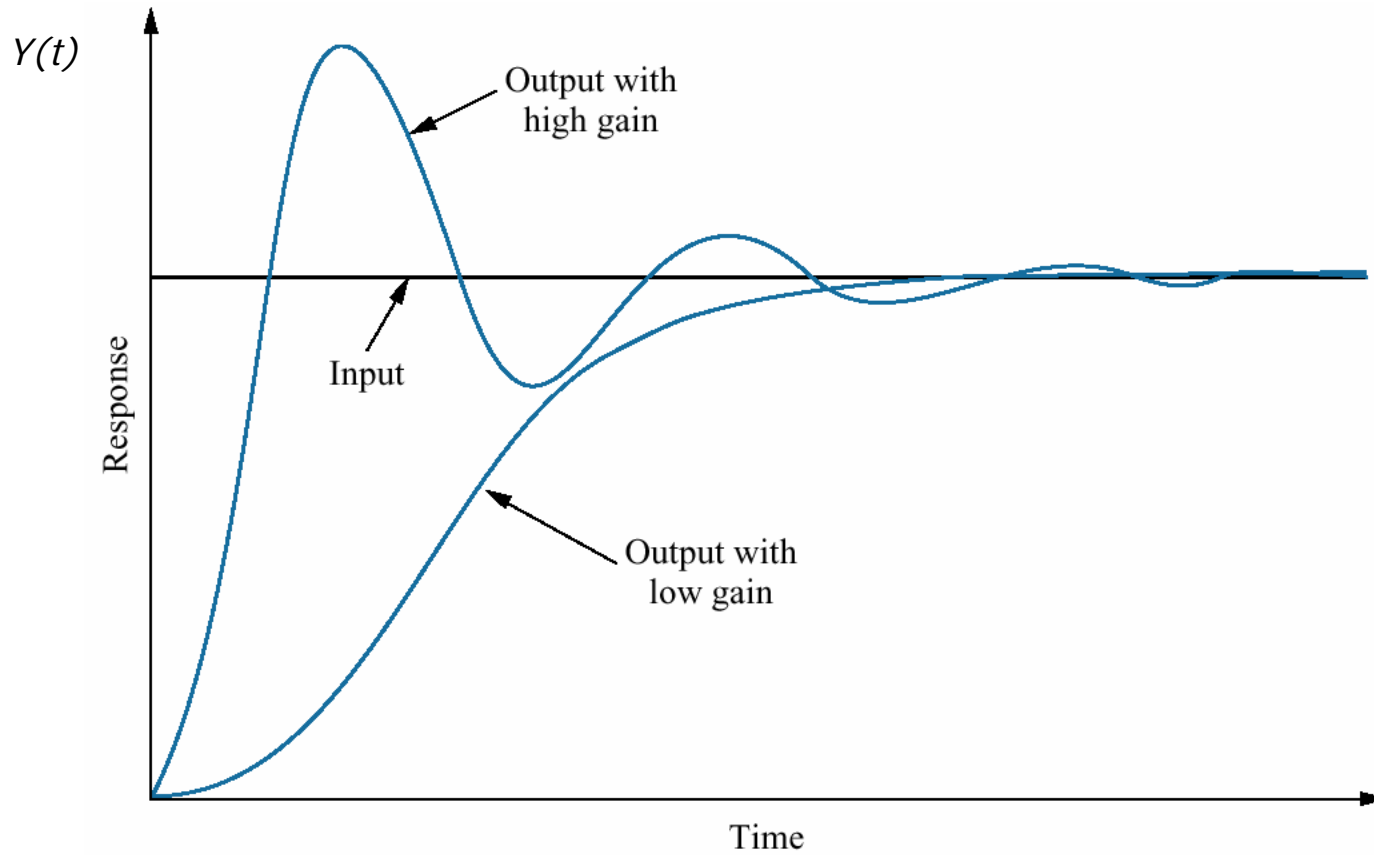


The underlying physical principles will lead to the formation of differential equation representation of the system as follows:

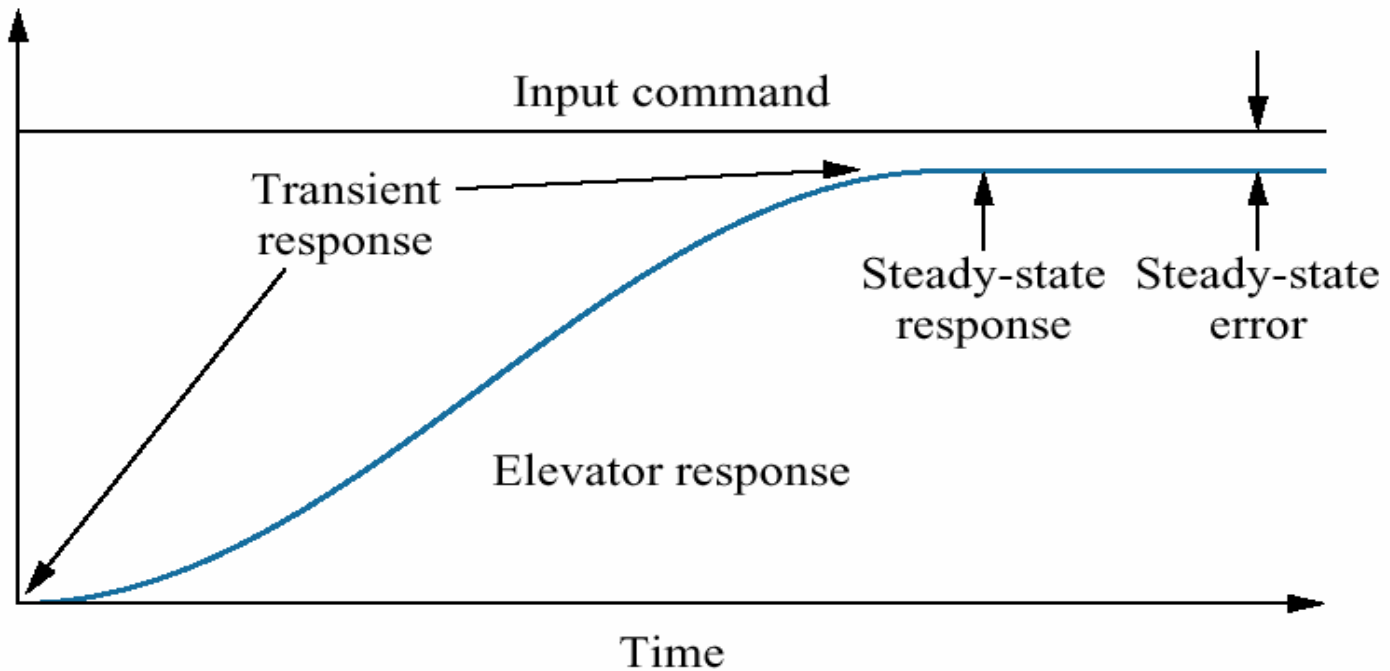
$$a_n \frac{d^n}{dt^n} y + a_{n-1} \frac{d^{n-1}}{dt^{n-1}} y + \dots + a_1 \frac{d}{dt} y + a_0 y = u(t)$$

The idea is to find out how system output,  $y(t)$ , responds to a particular type of input excitation,  $u(t)$ .

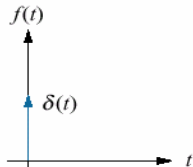
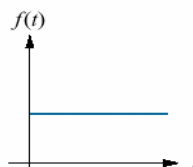
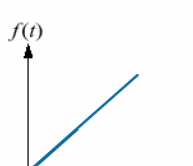
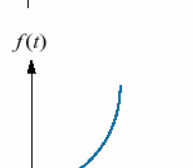
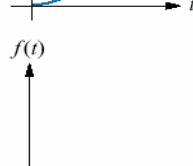
## Typical step response



## Some definition of terms

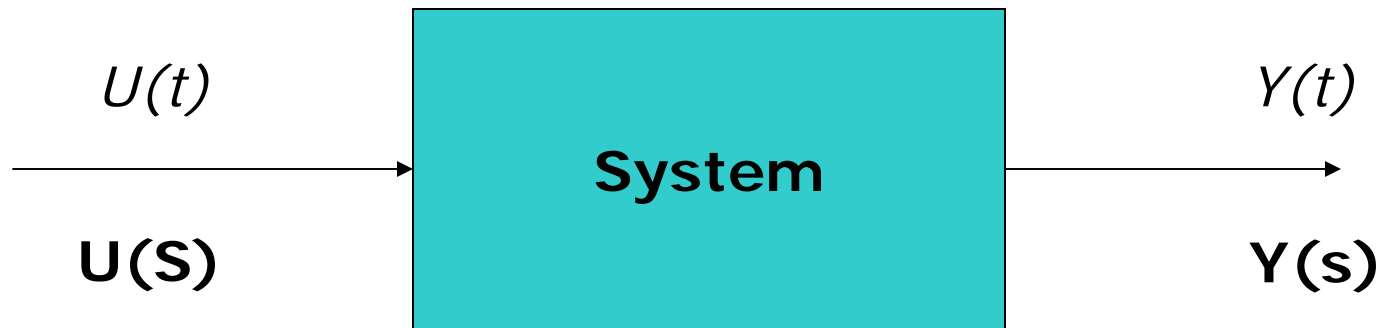


# Typical test inputs for control system analysis

Input	Function	Description	Sketch	Use
Impulse	$\delta(t)$	$\delta(t) = \infty$ for $0- < t < 0+$ $= 0$ elsewhere $\int_{0-}^{0+} \delta(t) dt = 1$		Transient response Modeling
Step	$u(t)$	$u(t) = 1$ for $t > 0$ $= 0$ for $t < 0$		Transient response Steady-state error
Ramp	$tu(t)$	$tu(t) = t$ for $t \geq 0$ $= 0$ elsewhere		Steady-state error
Parabola	$\frac{1}{2}t^2u(t)$	$\frac{1}{2}t^2u(t) = \frac{1}{2}t^2$ for $t \geq 0$ $= 0$ elsewhere		Steady-state error
Sinusoid	$\sin \omega t$			Transient response Modeling Steady-state error

## Transfer function representation of the system

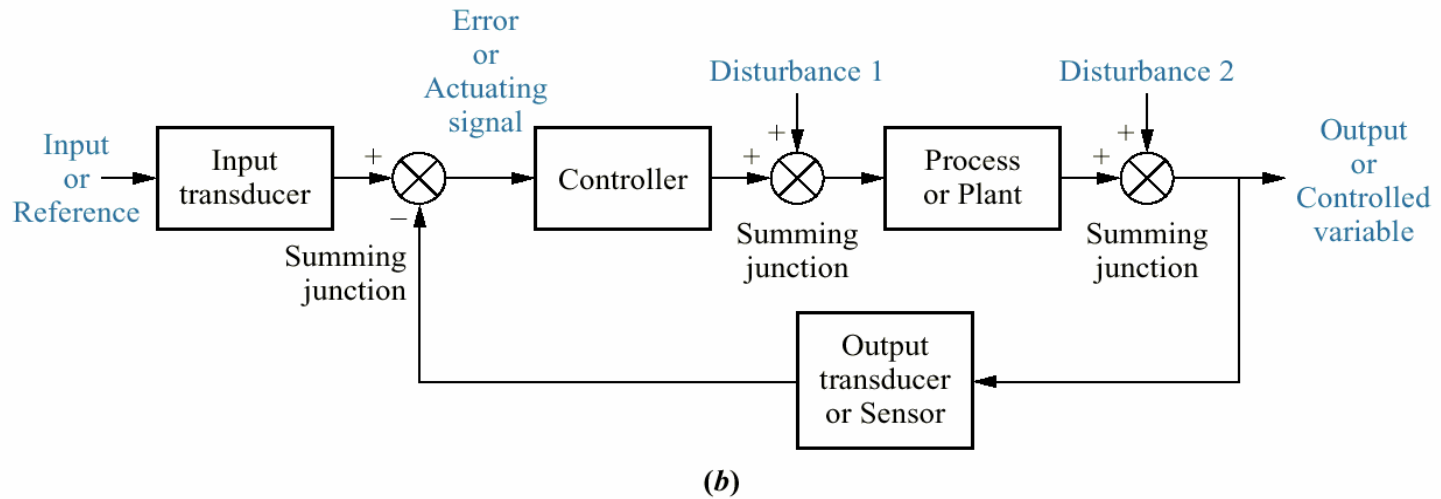
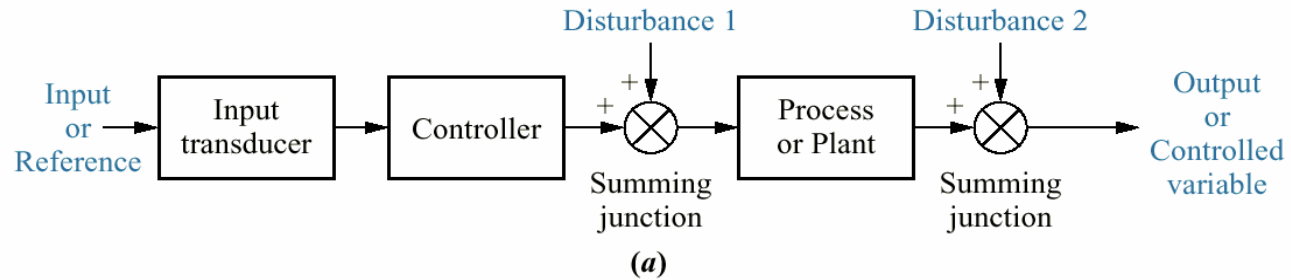
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Transfer Function:  $G(s) = Y(s)/U(s)$

*Under zero initial conditions*

# A typical control system



## State space representation of dynamic systems

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In addition to the transfer function representation, one can also represent a dynamic system in terms of state space form

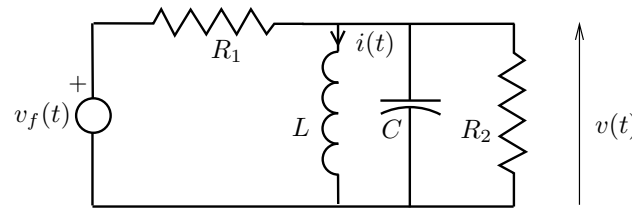
$$\frac{dx}{dt} = f(x(t), u(t), t)$$
$$y(t) = g(x(t), u(t), t)$$

If the system is linear, one can have the following representation:

$$\frac{dx(t)}{dt} = \mathbf{A}x(t) + \mathbf{B}u(t)$$
$$y(t) = \mathbf{C}x(t) + \mathbf{D}u(t)$$

## Example

Consider the simple electrical network shown below. Assume we want to derive the dynamic relationships between the input  $v_f(t)$  the voltage  $v(t)$



On applying fundamental network laws we obtain the following equations:

$$v(t) = L \frac{di(t)}{dt}$$

$$\frac{v_f(t) - v(t)}{R_1} = i(t) + C \frac{dv(t)}{dt} + \frac{v(t)}{R_2}$$

## Example (Cont)

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These equations can be rearranged as follows:

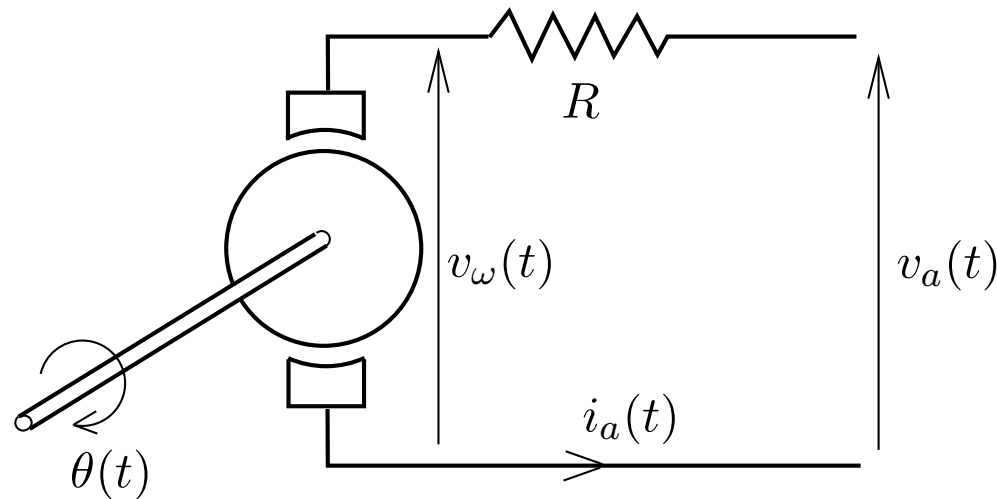
$$\frac{di(t)}{dt} = \frac{1}{L}v(t)$$
$$\frac{dv(t)}{dt} = -\frac{1}{C}i(t) - \left(\frac{1}{R_1C} + \frac{1}{R_2C}\right)v(t) + \frac{1}{R_1C}v_f(t)$$

We have a linear state space model with

$$\mathbf{A} = \begin{bmatrix} 0 & \frac{1}{L} \\ -\frac{1}{C} & -\left(\frac{1}{R_1C} + \frac{1}{R_2C}\right) \end{bmatrix}; \quad \mathbf{B} = \begin{bmatrix} 0 \\ \frac{1}{R_1C} \end{bmatrix}; \quad \mathbf{C} = [0 \quad 1]; \quad \mathbf{D} = \mathbf{0}$$

## Example of a DC motor

Consider a separately excited D.C. motor. Let  $v_a(t)$  denote the armature voltage,  $\theta(t)$  the output angle. A simplified schematic diagram of this system is shown below



## Parameters of DC motor

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Let

- $J$  - be the inertia of the shaft
- $\tau_e(t)$  - the electrical torque
- $i_a(t)$  - the armature current
- $k_1; k_2$  - constants
- $R$  - the armature resistance

Application of well known principles of physics tells us that the various variables are related by:

## Dynamic model of a DC motor

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$$J\ddot{\theta}(t) = \tau_e(t) = k_1 i_a(t)$$

$$v_\omega(t) = k_2 \dot{\theta}(t)$$

$$i_a(t) = \frac{v_a(t) - k_2 \dot{\theta}(t)}{R}$$

$$\frac{d}{dt} \begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & \frac{-k_1 k_2}{R} \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} + \begin{bmatrix} 0 \\ \frac{k_1}{R} \end{bmatrix} v_a(t)$$

## Summary of modeling of dynamic systems

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- In order to systematically design a controller for a particular system, one needs a formal - though possibly simple - description of the system. Such a description is called a model.
- A model is a set of mathematical equations that are intended to capture the effect of certain system variables on certain other system variables.

- *Certain system variables:* It is usually neither possible nor necessary to model the effect of every variable on every other variable; one therefore limits oneself to certain subsets. Typical examples include the effect of input on output, the effect of disturbances on output, the effect of a reference signal change on the control signal, or the effect of various unmeasured internal system variables on each other.

## Points to note

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- *Capture*: A model is never perfect and it is therefore always associated with a modeling error. The word capture highlights the existence of errors, but does not yet concern itself with the precise definition of their type and effect.
- *Intended*: This word is a reminder that one does not always succeed in finding a model with the desired accuracy and hence some iterative refinement may be needed.
- *Set of mathematical equations*: There are numerous ways of describing the system behavior, such as linear or nonlinear differential or difference equations.

- Models are classified according to properties of the equation they are based on. Examples of classification include:

<i>Model Attribute</i>	<i>Contrasting Attribute</i>	<i>Asserts whether or not ...</i>
Single input Single output	Multiple input multiple output	... the model equations have one input and one output only
Linear	Nonlinear	... the model equations are linear in the system variables
Time varying	Time invariant	... the model parameters are constant
Continuous	Sampled	... model equations describe the behavior at every instant of time, or only in discrete <i>samples</i> of time
Input-output	State space	... the model equations rely on functions of input and output variables only, or also include the so called <i>state variables</i> .
Lumped parameter	Distributed parameter	... the model equations are ordinary or partial differential equations